

## Extracting science from surveys of our Galaxy

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**Abstract.** Our knowledge of the Galaxy is being revolutionised by a series of photometric, spectroscopic and astrometric surveys. Already an enormous body of data is available from completed surveys, and data of ever increasing quality and richness will accrue at least until the end of this decade. To extract science from these surveys we need a class of models that can give probability density functions in the space of the observables of a survey – we should not attempt to “invert” the data from the space of observables into the physical space of the Galaxy. Currently just one class of model has the required capability, so-called “torus models”. A pilot application of torus models to understanding the structure of the Galaxy’s thin and thick discs has already produced two significant results: a major revision of our best estimate of the Sun’s velocity with respect to the Local Standard of Rest, and a successful prediction of the way in which the vertical velocity dispersion in the disc varies with distance from the Galactic plane.

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### 1. Introduction

When Chandra worked on stellar dynamics, the subject was largely concerned with understanding the phase-space distribution of near-by stars. Subsequently the subject’s focus shifted to understanding the large-scale dynamics of globular clusters and external galaxies, for which rich bodies of observational data started to be available in the 1970s. Recently there has been renewed interest in the dynamics of our Galaxy as a result of spectacular improvements in the quantity and quality of data for our Galaxy.

The richness of the available data, and the strong impact that observational selection effects have upon it, make the task of extracting science from Galaxy surveys qualitatively different from anything previously attempted in stellar dynamics. New methods in stellar dynamics, and new methods of data analysis, will have to be developed if we are to do justice to the data that will be available by the end of the decade. In this article I describe the challenge and describe one way of addressing it.

### 2. The current challenge

We are in the middle of the golden age of surveys of our Galaxy. These surveys are usefully divided into photometric surveys, spectroscopic surveys and astrometric surveys.

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Important near-infrared photometric surveys include the 2MASS and DENIS surveys of the infrared sky, which were completed in the last ten years, the UKIDSS survey, which is nearing completion and a couple of deeper surveys that are now getting underway with ESO's VISTA telescope. The SDSS project, which was completed a few years ago, obtained visual multi-band photometry of tens of millions of stars

The SDSS project was extended by the SEGUE project, and together these surveys obtained low-resolution spectra of several hundred thousand stars. The RAVE survey, which is nearing completion, will provide spectra at resolution  $R = 7\,500$  for  $\sim 500\,000$  stars. The LAMOST telescope, which is currently being commissioned, will take optical spectra of huge numbers of halo stars. The APOGEE project, which aims to gather 100 000 near-infrared spectra in the northern hemisphere at  $R = 30\,000$ , will soon be taking data. Meanwhile the HERMES project will obtain a similar number of near-infrared spectra in the southern hemisphere, and a large spectral survey with the VLT is to be undertaken by ESO.

Astrometric astronomy was revolutionised by the European Space Agency's Hipparcos mission, which published a catalogue of  $\sim 100\,000$  parallaxes in 1997. Hipparcos established an all-sky reference frame that was tied to quasars. The US Naval Observatory used this reference frame to re-reduce a large body of terrestrial observations, leading to the UCAC3 catalogue, which gives proper motions for  $10^8$  stars. The Pan-Starrs survey is beginning to image much of the sky to magnitude  $V \sim 24$  on a regular basis. It will discover enormous numbers of variable stars and measure parallaxes and proper motions for all the objects it detects. In 2013 ESA will launch Gaia, the follow-on to Hipparcos and the first satellite to conduct an astrometric survey of the sky – Hipparcos had an input catalogue while Gaia will itself identify objects. Gaia will obtain astrometry of unrivalled precision down to magnitude  $V \sim 20$  and spectra for objects brighter than  $V \sim 17$ . In all the Gaia Catalogue, which should be published around 2020, will contain astrometry for  $\sim 10^9$  stars and stellar parameters and line-of-sight velocities for  $\sim 10^8$  stars.

### **3. How predictive is $\Lambda$ CDM?**

The Cold Dark Matter model of cosmology, latterly with the modification that vacuum energy contributes an appreciable cosmological constant, has been a spectacular success in accounting for the pattern of fluctuations in the cosmic background radiation and the clustering of galaxies. In light of these successes it is often stated that the goal of surveys of the Milky Way is to “test  $\Lambda$ CDM”.

I think this point of view over-estimates the predictive power of the  $\Lambda$ CDM model, which is after all a theory of the invisible. Its successes are based on the ease with which the clustering of collisionless matter can be computed when it dominates the gravitational field. In particular, before the era of decoupling, the photon/baryon fluid was nearly homogeneous, and until the era of the first stars ( $z \simeq 15?$ ), the baryons were no more strongly clustered than the dark matter (DM) so the gravitational field was everywhere dominated by the DM distribution. Consequently, the distribution of DM can be quite reliably computed from the linear regime up to the era of the first stars.

From that time on the theory becomes enormously more intractable because crucial events were occurring in regions dominated by the gravitational field of the baryons, and that field depends on the tremendously complex physics of baryons: strong interactions (nucleosynthesis), weak interactions (supernova blast waves), and electromagnetic interactions (stellar winds and photo-heating to name but two key processes) all have a major impact of the distribution of matter and thus on the driving gravitational field.

For more than a decade a relatively small group of courageous theorists have been endeavouring to include key aspects of baryon physics in simulations of cosmic evolution. Although they are extremely complex and challenging, these simulations are not truly “ab-initio” in the sense that they take the well-understood physics of stars as read and merely try to follow how stars form, and how energy that is released by them impacts the surrounding interstellar medium. Truly ab-initio simulations require a dynamic range that is so large that it will probably never be possible, and we do not know what dynamic range has to be attained to make reliable simulations of the type that are currently being undertaken. Until that dynamic range has been attained, we do not know what predictions  $\Lambda$ CDM makes for the structures of galaxies.

The job of surveys of the Milky Way is to discover what is out there, and we should approach this job in an open-minded spirit.

#### **4. Discrete models are not enough**

N-body simulations have been of crucial importance for the development of our understanding of stellar dynamics. They have, moreover, been key to the development of the  $\Lambda$ CDM model, and irrespective of any reservations one may have about the extent to which model Galaxies extracted from cosmological simulations are based on irrefutable physics, these models have had a big impact on our thinking, and they will undoubtedly play a significant role in the future. Nevertheless, I will now argue that we need other types of models too.

The major issue with N-body models is the difficulty of fitting them to data. This problem has several aspects

- A high-quality model is computationally expensive to produce, and the connection between its structure and the initial conditions and phenomenological model from which it started is unclear. Consequently, it is hard to know how the initial conditions or input sub-grid physics should be modified to obtain a model that provides a better fit to given data. This objection has recently been weakened by the development of “made-to-measure” (M2M) modelling [14, 32]. M2M modelling allows one to steer an N-body model towards a better fit to given observational data, and it has already produced the best current model of the inner Galaxy [5].
- The precise arrangement of the model’s particles is of no interest (a short time later every particle will be somewhere else); the model’s content is the underlying probability density function (pdf) of which the current particle distribution is just a discrete realisation. The precise distribution of material in the Galaxy is likewise a discrete realisation of an underlying pdf. If we had the pdf of the model, we could evaluate the likelihood of the Galaxy given the model. But we don’t have this pdf, and asking whether two discrete realisations are consistent with a common pdf is hard. The obvious way to accomplish this task is to estimate the pdf of one of the two realisations by binning the particles, but we shall see in §6 below that this procedure is problematic in the case of the Galaxy.
- Our location within the Galaxy leads to our seeing right down the luminosity function in the immediate vicinity of the Sun, but being able to detect only the most luminous stars far away in the disc or at the Galactic Centre. Crucial information is carried by low-luminosity objects seen only locally (for example old white dwarfs encode the history of star formation), and luminous stars are obviously important

tracers of the Galaxy’s large-scale structure. So models need to encompass the entire range of luminosities. However, as [7] pointed out, it is not clear how an N-body model can fulfil this criterion: if each particle represents a star, most particles will have to represent low-luminosity stars, so they will be invisible anywhere but near the Sun, while if particles represent cohorts of stars, a particle that lies near the Sun will give rise to many low-luminosity stars, all with identical kinematics. Again the essential point is that we really need the underlying pdf, not a discrete realisation of it.

I conclude that N-body models are not suited to the extraction of science from survey data because they lack flexibility and do not provide the required pdfs.

## 5. Steady-state models are crucial

The Galaxy is not in a steady state, most obviously because it contains a bar near the Centre, and spiral structure within the disc, and more subtly because the SDSS survey revealed that the stellar halo is largely comprised of streams and still-enigmatic “clouds” like the Hercules-Aquila Cloud [1]. None the less, modellers of the Galaxy have no option but to start by constructing steady-state models.

The reason for this necessity is that DM makes a major contribution to the Galaxy’s gravitational field – our best guess is that the Sun lies close to the radius at which baryon-domination at small radii gives way to DM-domination at large radii. At present, we can detect DM only through its contribution to the gravitational field, which we map through the influence it has on objects that we can see – on Galactic scales this amounts to studying the dynamics of gas and stars. In principle *any* phase-space distribution of stars is consistent with *any* gravitational field. It is only when we insist on some sort of statistical equilibrium that the distribution of stars imposes constraints on the gravitational field, and thus on the combined density of stars, gas and DM. For example, the assumption of statistical equilibrium enables us to rule out a weak gravitational field, because in that field the observed distribution of stars would expand systematically, while a very strong gravitational field can be excluded because it would cause the observed distribution of stars to contract systematically.

The argument just given is rather crude in that it does not engage with the more subtle devices for determining the gravitational field, such as the use of hydrodynamical simulations of the flow of gas [12, 35], or stellar streams [13, 21]. However, I believe these techniques also rest on the assumption that the Galaxy is broadly in statistical equilibrium.

Actually, even if we could observe DM directly, and hence determine the Galaxy’s gravitational field without recourse to dynamics, it would still make sense to seek an equilibrium model of the Galaxy first. Then comparing the predictions of this model with the data we would identify features that signalled departures from equilibrium, and we could seek to model these features by perturbing our equilibrium model. In this connection it is salutary to recall the extent to which the language of physics, and thus our understanding of phenomena, has been moulded by perturbation theory: dispersion relations, photons, phonons, Feynman diagrams, orbital elements, mean-motion resonances, etc., are all concepts, introduced by perturbation theory, that loom large in our understanding of how the world works. Thus they are both useful mathematical abstractions and essential tools for understanding. Historically, perturbation theory has been rather little used in galactic dynamics, and we are all the poorer in understanding for it.

## 6. We need to fit models to data in data space

We like to conceive of the Galaxy as an object that lives either in three-dimensional “real” space, or better in six-dimensional phase space. Actually, below I shall argue that even the most basic Galaxy model inhabits a space of at least ten dimensions, but for the moment let’s be conservative and imagine that it inhabits phase space, where we can readily write down the equations that govern the evolution in time of the pdf of a system of mutually gravitating particles.

Unfortunately, we do not directly measure the natural coordinates of this space. For example, Gaia will measure two angular coordinates  $(\alpha, \delta)$  and a parallax  $\varpi$  instead of a distance. For many stars it will measure the line of sight velocity  $v_{\text{los}}$  and for all objects it will measure two components of the proper motion  $\mu$ . So for many stars Gaia will measure six coordinates that form rather an odd set from the perspective of physics. In particular, the star’s physical location and two of its components of velocity depend on the measured parallax, which for a distant star may be measured to be negative. Clearly a negative parallax is unphysical, but it does carry information: it tells us that the star is more distant than the distance that corresponds to the uncertainty in the parallax. Our modelling strategy must be such that we can make good use of stars with negative measured parallaxes

Because the inverse of a negative parallax cannot be interpreted as a distance, a strategy that is *not* going to work is to infer the star’s phase-space coordinates from the data; we cannot carry the star from the space of the data into the space of the model, namely phase space. We *must* proceed in the opposite direction, projecting the model into the space of the observables. When we carry the model into the space of observables, negative parallaxes are in no way problematic – indeed a catalogue in which negative parallaxes did not occur is what would be problematic.

Even for stars with safely positive parallaxes, there is a huge advantage in carrying the model into the space of observables because in this space the errors are likely to be largely uncorrelated and even Gaussian. Indeed, they will be the result of a large number of statistically independent chance events: the displacement of the measured photocentre from its true location is determined by photon-counting statistics, distortions of the CCD’s grid, etc. The pdf in phase space into which a Gaussian pdf in  $(\alpha, \delta, \varpi, v_{\text{los}}, \mu_\alpha, \mu_\delta)$  space would map would be strongly non-Gaussian and predict large correlations between errors in distance and the components of tangential velocity, for example. With such a complex pdf for the errors, we have no hope of achieving a satisfactory understanding of how errors give rise to uncertainties in the parameters of our models.

Given that we must carry our models into the space of observables, let us examine more carefully what coordinates that space really has. In addition to  $(\alpha, \delta, \varpi, v_{\text{los}}, \mu_\alpha, \mu_\delta)$  we will invariably measure an apparent magnitude  $m$ , a colour such as  $V - I$ . The spectrum from which  $v_{\text{los}}$  was extracted will also have yielded the star’s surface gravity  $\log g$  and measures of metallicity, such as  $[\text{Fe}/\text{H}]$  and possibly  $[\alpha/\text{Fe}]$ . If a medium- or high-resolution spectrum is available, there will be measurements of the abundances of many individual chemical elements, such as C, O, Mg, Ti, Eu, etc. Hence, the dimensionality of the space of observables  $(\alpha, \delta, \varpi, v_{\text{los}}, \mu_\alpha, \mu_\delta, m, V - I, \log g, [\text{Fe}/\text{H}], \dots)$  will normally be as high as 10, and it may be significantly higher. Our modelling strategy must be designed to cope with such large dimensions.

The high dimensionality of the space of data makes any analysis that involves binning the data very unattractive. To see this, imagine establishing a Cartesian grid in data space, the cell spacing in most dimensions being of order half the typical observational uncertainty of the corresponding observable. If we applied this criterion to  $l$  and  $b$ , we would

obtain an absurdly fine grid on the sky. Therefore, in  $l$  and  $b$  we take the separation between bins to be comparable to the changes in angle over which we expect the distribution of stars to change significantly – this might be as large as  $10^\circ$  in  $l$  and a degree or so in  $b$  at low latitudes and larger increments near the poles. Having established our grid, we assign stars to cells and thus obtain the density of stars in each bin. The Poisson noise in our density estimates decreases with the number of stars assigned to each cell, which increases with the adopted bin sizes and decreases with the number of quantities actually observed. Hence binning is most attractive for the sparsest data set, namely measurements of  $(\alpha, \delta, m, \mu_\alpha, \mu_\delta)$ . However, this is already a five-dimensional space. For a survey of a third of the sky, it might be possible to get by with 100 bins on the sky. A few tens of bins would be required for the apparent magnitudes, and for each component of  $\mu$  ten bins might suffice. Thus a minimal grid will have  $> 10^5$  bins, so even with the simplest conceivable data, binning first becomes advantageous when the number of stars in the catalogue exceeds a million. In the case of Gaia, the list of observables must be expanded to include at least  $\varpi$ , and the number of bins required to do justice to the proper-motion data must be increased by a factor of at least 10, implying a grid with  $> 10^6$  bins. In reality one would want to include colour data, and some line-of-sight velocities, and the number of bins would be pushed up to  $\sim 10^9$ , essentially the number of stars in the catalogue. Thus we should not rely on binning the data.

## 7. We must use multiple lines of evidence

If a star is near enough, no measure of distance can trump a parallax. But usually the majority of the objects measured in a survey lie far away, and then a parallax measurement carries less information. For these objects spectrophotometric distances are likely to be important. In many cases a small parallax will imply that the star is distant, and from its apparent magnitude it will be evident that it is a giant. This information can inform the choice of template star used in the analysis of the star's spectrum and thus the determination of its values of  $v_{\text{los}}$ ,  $\log g$  and  $[\text{Fe}/\text{H}]$ , as well as its distance. Thus the pdf of the distance upon which we finally settle will depend on several sources of information and a great deal of modelling.

Several of the quantities we measure are connected by well-understood physics. For example, the theory of stellar evolution constrains stars to a small subset of  $(T_{\text{eff}}, \log g, [\text{Fe}/\text{H}])$  space. The principles of Bayesian inference give us a framework for using such constraints to reduce the uncertainties in stellar parameters [e.g. 9]. Ideally we would extract stellar parameters in parallel with fitting a Galaxy model to the data, since the stellar parameters depend on the distance, just as the parallax and tangential velocities do. This scheme appears to be hard to implement in practice, so in the near future stellar parameters extracted from spectrophotometric data in isolation will play a large role in Galaxy modelling. We should not lose sight, however, of the long-term goal of expanding the “observables” to include the calibrated spectrum and fitting the model to the data in this expanded space.

## 8. Torus models

I believe only one class of models fulfils the requirements we have identified above, namely, of being steady-state models that yield a pdf in the space of observables, and can be used as a basis for studying non-equilibrium features by means of perturbation theory. This is the class of “torus models”, which I will now describe.

Torus modelling is essentially a variant of Schwarzschild modelling [29], which has been for more than a decade the standard tool with which to extract dynamical information from observations of external galaxies (especially in connection with searches for massive central black holes [16]). To build a Schwarzschild model one guesses the form of the system's gravitational potential and then integrates a representative “library” of orbits in the chosen potential. Finally one seeks non-negative weights for each orbit such that the weighted sum of the contributions of each orbit to the observables is consistent with the data. If we can find no satisfactory solution with the current gravitational potential  $\Phi$ , then we adjust  $\Phi$  and try again.

Torus modelling involves replacing orbits, which are time series of phase-space points, by orbital tori, which are analytic expressions for the three-dimensional continuum of phase-space points that are accessible to a star on a given orbit. Whereas an orbit is labelled by its initial conditions, a torus is labelled by its three action integrals  $J_i$ ; whereas position on an orbit is determined by the time  $t$  elapsed since the initial condition, position on a torus is determined by the values of three angle variables  $\theta_i$ , one canonically conjugate to each action  $J_i$ . For a summary of how orbital tori are constructed and references to the papers in which torus dynamics was developed, see [26].

Replacing numerically integrated orbits with orbital tori brings the following advantages:

- The density of stars in phase space is simply related to the sampling density of tori in action space and the weights assigned to those tori. This is because tori have prescribed actions and the six-dimensional phase-space volume occupied by orbits with actions in  $d^3\mathbf{J}$  is  $\tau = (2\pi)^3 d^3\mathbf{J}$ . Knowledge of the phase-space density of orbits allows one to convert between orbital weights and the value taken by the DF on an orbit. It proves advantageous to choose an analytic representation of the DF  $f(\mathbf{J})$  and to derive the weights of individual tori from  $f$ . The weights are varied by adjusting parameters that appear in  $f$  [2].
- There is a clean and unambiguous procedure for sampling orbit space. The choice of initial conditions from which to integrate orbits for a library is less straightforward because the same orbit can be started from many initial conditions, and when the initial conditions are systematically advanced through six-dimensional phase space, the resulting orbits are likely at some point to cease exploring a new region of orbit space and start resampling a part of orbit space that is already represented in the library. On account of this effect, it is hard to relate the weight of an orbit to the value taken by the DF on it [but see 19, 33, for how this can be done].
- There is a simple relationship between the distribution of stars in action space and the observable structure and kinematics of the model – as explained in §4.6 of [4], the observable properties of a model change in a readily understood way when stars are moved within action space. The simple relationship between the observables and the distribution of stars in action space enables us to infer from the observables the approximate form of the DF  $f(\mathbf{J})$ , which is nothing but the density of stars in action space.
- From a torus one can readily find the velocities that a star on a given orbit can have when it reaches a given spatial point  $\mathbf{x}$ . By contrast a finite time series of an orbit is unlikely to exactly reach  $\mathbf{x}$ , and searching for the time at which the series comes closest to  $\mathbf{x}$  is laborious. Moreover, several velocities are usually possible at a given location, and a representative point of closest approach must be found for each possible velocity.

- An orbital torus is represented by of order 100 numbers while a numerically-integrated orbit is represented either by some thousands of six-dimensional phase-space locations, or by a similar number of occupation probabilities within a phase-space grid.
- The numbers that characterise a torus are smooth functions of the actions  $\mathbf{J}$ . Consequently tori for actions that lie between the points of any action-space grid can be constructed by interpolation on the grid. Interpolation between time series is not practicable.
- Schwarzschild and torus models are zeroth-order, time-independent models which differ from real galaxies by suppressing time-dependent structure, such as ripples around early-type galaxies [25, 27, 30], and spiral structure or warps in disc galaxies. Since the starting point for perturbation theory is action-angle variables [e.g. 23], in the case of a torus model one is well placed to add time-dependent structure as a perturbation. [22] showed that classical perturbation theory works extremely well when applied to torus models because the integrable Hamiltonian that one is perturbing is typically much closer to the true Hamiltonian than in classical applications of perturbation theory [11, 17, 34], in which the unperturbed Hamiltonian arises from a potential that is separable (it is generally either spherical or plane-parallel).

The impact of shot noise on the model is usually minimised if all tori have the same weight, and this will be the case if the density of used tori in action space samples the DF. We endeavour to ensure that this condition is met, at least to a good approximation [3].

## 9. The interface between the thin and thick discs

A considerable body of evidents now points to the disc of our Galaxy being a superposition of a “thick disc” made up of old stars on orbits that are moderately eccentric and inclined to the Galactic plane, and a thin disc of stars with higher ratios of Fe to O and Mg abundances that are on less strongly eccentric or inclined orbits. The details of the superposition are, however, murky because stars from both populations are found near the Sun and even at the same velocities.

[2] proposed assigning a simple DF to each population. For the thick disc he proposed

$$f_{\text{thk}}(J_r, J_z, L_z) = f_{\sigma_r}(J_r, L_z) f_{\sigma_z}(J_z), \quad (1)$$

where  $f_{\sigma_z}$  is defined by

$$f_{\sigma_z}(J_z) \equiv \frac{e^{-\Omega_z J_z / \sigma_z^2}}{2\pi \int_0^\infty dJ_z e^{-\Omega_z J_z / \sigma_z^2}}. \quad (2)$$

Here  $\Omega_z(\mathbf{J})$  is the fundamental frequency of vertical oscillations,  $\sigma_z$  is a constant with the dimensions of velocity, and the denominator ensures that  $f_{\sigma_z}$  satisfies the normalisation condition

$$\int dz dv_z f_{\sigma_z} = 1 \quad \Leftrightarrow \quad \int dJ_z f_{\sigma_z} = \frac{1}{2\pi}. \quad (3)$$

Similarly,  $f_{\sigma_r}$  is defined by

$$f_{\sigma_r}(J_r, L_z) \equiv \frac{\Omega \Sigma}{\pi \sigma_r^2 \kappa} \bigg|_{R_c} [1 + \tanh(L_z / L_0)] e^{-\kappa J_r / \sigma_r^2}. \quad (4)$$



Here  $R_c(L_z)$  is the radius of the circular orbit with angular momentum  $L_z$ ,

$$\Sigma(L_z) = \Sigma_0 e^{(R_0 - R_c)/R_d} \quad (5)$$

is the thick disc's surface density, and  $\Omega(L_z)$  and  $\kappa(L_z)$  are the circular and the radial epicycle frequencies there. With these choices the disc's surface density is approximately exponential in  $R$  with scale length  $R_d$ . We take  $L_0 \ll R_0 v_c(R_0)$  so the term in square brackets serves to eliminate retrograde stars.

The scale heights of the discs of external galaxies seem to be approximately independent of radius, and this finding suggests that the vertical velocity dispersion in these discs is roughly proportional to  $e^{-R/2R_d}$ . We enforce similar behaviour on the radial and vertical dispersions within the thick disc by taking both  $\sigma_r$  and  $\sigma_z$  to be functions of  $L_z$ :

$$\begin{aligned} \sigma_r(L_z) &= \sigma_{r0} e^{q(R_0 - R_c)/R_d} \\ \sigma_z(L_z) &= \sigma_{z0} e^{q(R_0 - R_c)/R_d}. \end{aligned} \quad (6)$$

The only important parameters of  $f_{\text{thk}}$  are  $\sigma_{r0}$  and  $\sigma_{z0}$ .

For the DF of the thin disc, [2] proposed a superposition of “pseudo-isothermal” DFs like (1)

$$f_{\text{thn}}(J_r, J_z, L_z) = \frac{\int_0^{\tau_m} d\tau e^{\tau/t_0} f_{\sigma_r}(J_r, L_z) f_{\sigma_z}(J_z)}{t_0(e^{\tau_m/t_0} - 1)}. \quad (7)$$

The idea here is that in the thin disc star formation has continued throughout the lifetime of the Galaxy at a rate that has declined exponentially with time constant  $t_0 \simeq 3$  Gyr. Throughout the lifetime of a cohort of coeval stars, scattering of its stars by non-axisymmetric fluctuations in the gravitational potential has increased its velocity dispersions roughly as a power law of age,  $\sigma \propto \tau^\beta$ , so in equation (7) the dispersion parameters are given by

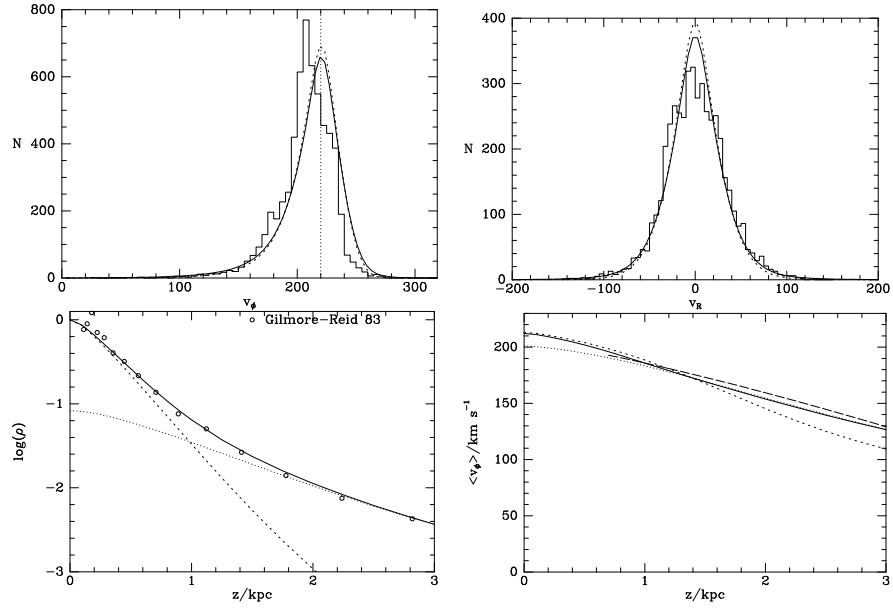
$$\begin{aligned} \sigma_r(L_z, \tau) &= \sigma_{r0} \left( \frac{\tau + \tau_1}{\tau_m + \tau_1} \right)^\beta e^{q(R_0 - R_c)/R_d} \\ \sigma_z(L_z, \tau) &= \sigma_{z0} \left( \frac{\tau + \tau_1}{\tau_m + \tau_1} \right)^\beta e^{q(R_0 - R_c)/R_d}, \end{aligned} \quad (8)$$

where  $\sigma_{r0}$  and  $\sigma_{z0}$  are constants.

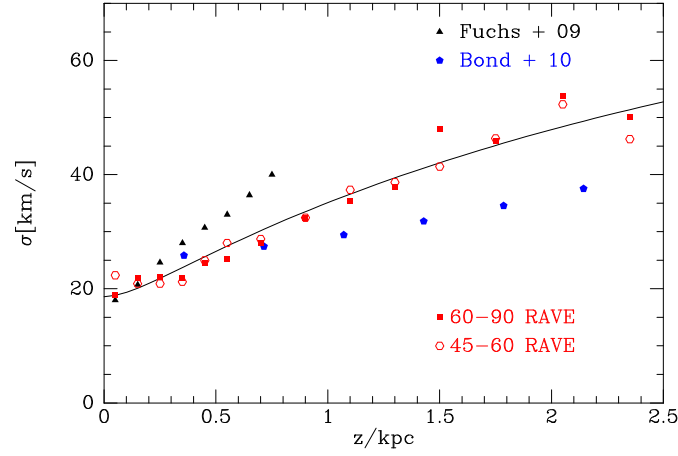
Fig. 1 shows prediction of the complete DF for the structure of the solar neighbourhood when  $\Phi_z(z)$  is taken to be the potential above the Sun in Model II of §2.7 in [4]; this model is disc-dominated. Full curves are for the whole disc and dashed curves show the contribution of the thin disc. The upper panels show for stars seen in the plane the distributions in  $v_\phi$  and  $v_R$  after integrating over the other two velocity components; the smooth curves are the predictions of the model, while the histograms show the distributions observed in the Copenhagen–Geneva survey [20]. The lower left panel shows that the model provides a good fit to the vertical density profile reported by [18], who discovered the thick disc. The lower-right panel shows the prediction of the model for how the mean-streaming speed should decrease with height above the plane.

### 9.1 The Sun's azimuthal velocity

The major weakness of the fits between model and data displayed in Fig. 1 is the displacement in the top left panel of the model curve to higher values of  $v_\phi$  than the histogram. It proves impossible to rectify this problem by changing the DF.



**Figure 1.** Structure at the solar radius predicted by the standard DF (eqs. 7 and 1). Full curves are for the entire disc while dashed curves show the contribution of the thin disc. The upper panels show velocity distributions for  $z = 0$  marginalised over the other velocity components. The vertical dotted line marks the local circular speed. In the lower panels dotted curves show the contributions of the thick disc. In the bottom-right panel the long-dashed line is the empirical fitting function of Ivezić et al. (2008). The values of the DF's parameters are given in Table 1.



**Figure 2.** The variation of  $\sigma_z$  with distance from the Galactic plane. The triangles and filled circles show conflicting determinations from the SDSS survey. The full curve is the model of [2]. The red squares and circles show the values subsequently determined from the RAVE survey by [8].

The histogram is here plotted as a function of  $v_\phi$ , the azimuthal velocity in the Galactic rest frame, whereas the data derive from measurements of velocities relative to the Sun. Consequently, the histogram can be moved to the right if we increase our estimate of the Sun's velocity relative to the local standard of rest. From Hipparcos data [10] determined this to be  $5.25 \pm 0.62 \text{ km s}^{-1}$  by plotting the mean motion relative to the Sun of different groups of stars, versus the square of the group's random velocity. A naive reading of Stromberg's equation [eq. (4.228) in 4] predicts that this plot will be linear §4.8.2(a) of [4], and the Hipparcos data showed that it was apart from the two points with the lowest random velocities, which must be affected by inadequate phase mixing of freshly formed stars. What [10] overlooked is that by grouping stars by colour they introduced a tendency for the groups with blue colours and low velocity dispersions to contain metal-poor stars, and vice versa for the groups of red stars. On account of the metallicity gradient in the disc, the guiding-centres of metal-poor stars tend to be at larger radii than those of metal-rich stars, so when they are visiting the solar neighbourhood, metal-poor stars tend to have larger  $v_\phi$  than metal-rich stars. That is, the metallicity gradient modifies the relation between colour and  $v_\phi$  that [10] used, and thus influenced their value of  $V_\odot$ . When [28] fitted an updated form of the Hipparcos data to a model of the chemodynamical evolution of the disc, they found  $V_\odot = 12 \pm 2 \text{ km s}^{-1}$ ,  $11\sigma$  larger than the old value. When the smooth curve in the top left panel of Fig. 1 is shifted to the right in accordance with the new value of  $V_\odot$ , it agrees with the data quite nicely, and in fact [2] proposed  $V_\odot = 11 \text{ km s}^{-1}$  for just this reason.

## 9.2 Dispersion in $W$ above the Sun

A second nice example of the value of models of the disc that have distribution functions analytic in  $\mathbf{J}$  is shown in Fig. 2. The lowest and highest series of points show estimates of  $\sigma_z(z)$  extracted from the SDSS by [6] and [15], respectively. These two estimates are clearly incompatible with each other. The smooth curve shows the run of  $\sigma_z$  with  $z$  in a model of the disc that [2] obtained by fitting the data for  $\rho(z)$  plotted in Fig. 1. [2] shows that this model also provides a good fit to the distribution of  $v_z$  of the GCS stars, although

this fit was not used to determine the model’s parameters. He was unable to find a model that was consistent with both these data and either the data sets derived from the SDSS. The red squares and circles in Fig. 2 show estimates of  $\sigma_z(z)$  that were subsequently extracted from the RAVE survey [31] by [8]: the squares use stars at  $b > 60^\circ$ , while the circles use stars at  $45^\circ < b < 60^\circ$ . The agreement between the two sets of points is excellent, which inspires confidence in the spectrophotometric distances to the RAVE stars that [8] used. The new data points also agree very well with the model of [2]. Thus we have a second example of data that proved impossible to fit using an analytic DF turning out to be themselves flawed, and the predictions of the DF being vindicated.

## 10. Conclusion

We are in the middle of a golden age for surveys of our Galaxy. We already have an enormous body of data, and data of ever more spectacular quality will continue to become available until at least the end of this decade. Extracting from these data an understanding of the current structure and the history of the Milky Way is going to be a formidable challenge. The sheer quantity and heterogeneity of the data make the task hard. The task is made harder still by our peculiar position within the Galaxy itself, as a consequence of which all surveys contain strong selection effects. Worse still, one of the crucial observables, parallax, can be easily scattered by measuring errors into negative values, which have no physical interpretation.

In light of these difficulties our strategy must be to build models from which we can compute the pdf of the observed quantities and to use this pdf to compute the likelihood of the data given the model. From this likelihood we can compute the pdf of the model’s parameters.

While N-body models are a key tool for developing our understanding of both galactic dynamics and cosmology, they appear to be unsuited to the interpretation of Galactic surveys because they do not deliver pdfs and they are hard to steer towards a model that is consistent with the data. A more promising way forward is offered by torus models, which are analogous to Schwarzschild models except that orbits (time-series of phase-space points) are replaced by orbital tori (three-dimensional subspaces of phase space). In principle the weights of individual tori can be independently assigned, as are the weights of orbits in a conventional Schwarzschild model, but a better strategy is to derive the weights from an analytic DF  $f(\mathbf{J})$ , and to vary these weights by varying a relatively small number of parameters in  $f$ .

The effectiveness of this strategy is demonstrated by using it to construct DFs for the thin and thick discs and adjusting its parameters to fit data from the Hipparcos, RAVE and SDSS surveys. This exercise led to the uncovering of a subtle error in the determination of the Sun’s velocity with respect to the local standard of rest, and led to the successful prediction of the variation of  $\sigma_z$  with distance  $z$  above the Galactic plane.

While this early work shows great promise, it does not conform to the methodology advocated here in that it fits the model to “data”, such as the number of stars in a range of values of  $v_\phi$ , that are in reality the result of carrying the data from the space of observables to the physical space to yield  $\bar{v}_\phi$ ,  $\rho(z)$  and  $\sigma_z(z)$ , for example. We are currently testing our ability to recover the DF by fitting to the data in the space of measured quantities by fitting DFs to pseudo-data constructed from similar DFs. First results are extremely encouraging in that they show that with only 10 000 stars the parameters of the DF can be recovered to good accuracy even when the stars are drawn from the (broad) general luminosity function, and the only data available are sky positions, proper motions and

apparent magnitudes. Complementing the data with measurements of parallax or line-of-sight velocity reduces the already small errors and especially correlations between the errors in the model's parameters.

There is a great deal of work to do in this field, but the outlook is exciting.

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